

Python and R Project Code:

Ritai Na

1. Introduction:

Shown below is a brief Introduction to Wind Turbine Status Analysis Project in the course **Machine Learning and Big Data:**

You may check the code in Github link Below.

Github Project Link: <https://github.com/BlueFamous/Machine-Learning-Course-Project>

2. Data Preprocessing:

(1) Dimension Reduction:

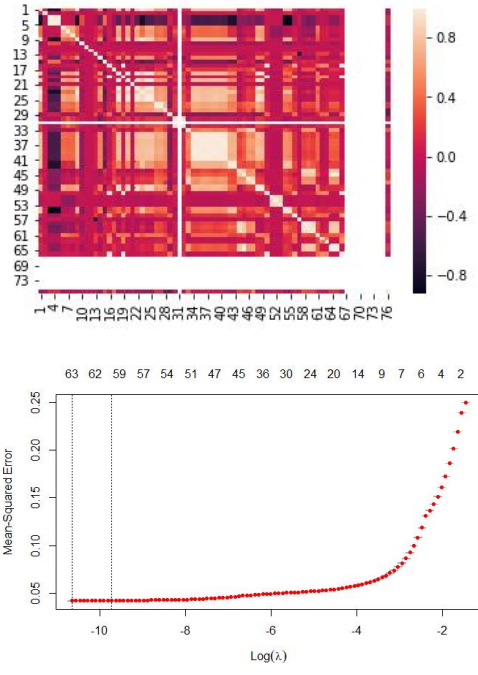
The original dataset consist of over 67 attributes and the total lines of data is over 1 million. To increase efficiency, I applied Lasso to make dimension regression using R.

(2) Oversampling to settle sample imbalance:

The original dataset is imbalanced in labels. Hence, I applied Borderline-SMOTE to settle this issue. The code and the result is shown below.

(3) Two-stage Transfer Adaboost:

Applied two-stage transfer adaboost to predict the labels.

Step	Original Code	Result
Dimension Reduction	<pre>```{r cars} library(openxlsx) library(glmnet) library(foreign) bc <- read.csv("C:\\Users\\HP\\Desktop\\01212.csv",header = FALSE) bc <- na.omit(bc) y<-as.matrix(bc[,49]) x<-as.matrix(bc[,c(2:48)]) f1 = glmnet(x, y, family="binomial", nlambda=100, alpha=1) plot(f1, xvar="lambda", label=TRUE) cvfit=cv.glmnet(x,y) plot(cvfit) cvfit\$lambda.min#Find the mean value cvfit\$lambda.1se#Find the lamda within one standard diviation ``` ```{r cars2} l.coef2<-coef(cvfit\$glmnet.fit,s=0.00037329,exact = F) l.coef1<-coef(cvfit\$glmnet.fit,s=0.000289,exact = F) l.coef1 l.coef2 ``` ```{r cars} library(openxlsx) library(glmnet) library(foreign) bc <- read.csv("C:\\Users\\HP\\Desktop\\0120.csv",header = TRUE) bc <- na.omit(bc) y<-as.matrix(bc[,1]) y ```</pre>	

Overs
ampli
ng

```
import pandas as pd
io = r'C:\Users\HP\Desktop\551.xlsx'
data = pd.read_excel(io, sheet_name=0, header=None)
import numpy as np

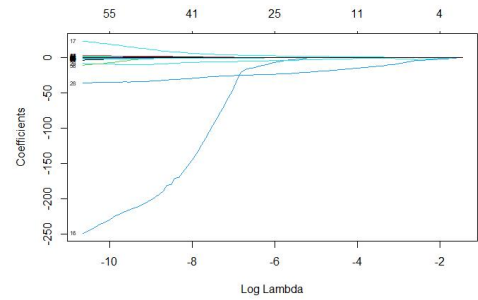
#
X = data.ix[:,0:39].values # Variables
y = data.ix[:,39].values # Dependent
##'''Using Borderline-SMOTE to settle sample imbalance'''
#from imblearn_under_sampling import ClusterCentroids
#cc = ClusterCentroids(random_state=0)
#X_resampled, y_resampled = cc.fit_sample(X, y)

#from imblearn_over_sampling import RandomOverSampler

#ros = RandomOverSampler(random_state=0)
#X_resampled, y_resampled = ros.fit_sample(X, y)

from imblearn_over_sampling import SMOTE
X_resampled, y_resampled = SMOTE().fit_resample(X, y)
print(X_resampled)
print(y_resampled)
# Integration of Data
data_resampled = np.zeros([len(X_resampled[:,0]),40])
data_resampled[:,40] = X_resampled
data_resampled[:,39] = y_resampled

data_resampled2 = pd.DataFrame(data_resampled)
writer = pd.ExcelWriter(r'C:\Users\HP\Desktop\999.xlsx')
data_resampled2.to_excel(writer)
writer.save()
writer.close()
```



Two-stage
Trade
boost
Code

```
import pandas as pd
import numpy as np
from sklearn import svm
from sklearn.model_selection import train_test_split
from sklearn.metrics import roc_auc_score
from sklearn import tree
import os
from sklearn import svm

path = "D:\\\\..."

def testDataset(train_x, label_x, test_x, test_y):
    # train_label = np.concatenate((label_x, label_y), axis=0)
    # Calculate number of rows
    row_x = train_x.shape[0]
    row_y = test_x.shape[0]
    # Append train data to shape, test shape
    test_data = np.concatenate((train_data, test), axis=0)

    # Weight Initialization
    weights_A = np.ones((row_x, 1)) /
        row_x # row number is A, col number is 1
    weights_B = np.ones((row_y, 1)) / row_y
    # Append test data to weights_A, weights_S, axis=0
    #data = np.concatenate((weights_A, weights_S), axis=0)
    data = 1 / (1 + np.exp(-(x * np.log(row_x / row_y)))
    # Store the labels and data
    test_A = np.zeros((row_y, 1))
    result_label = np.ones((row_x + row_y, 1))

    # predict = np.zeros([row_y])
    # Initial finish
    # row array
    # train_data = np.asarray(train_data, order='C')
    # test_label = np.asarray(test_label, order='C')
    # test_data = np.asarray(test_data, order='C')
```

[illegible]

```

% error_rate = 0.001
bata_T0_1 = error_rate / ( 1 - error_rate)
% Adjust the sample weights
ZT = ones(1, numSamples) + (1 + 1) *
    ( N - 1 ) * ( 1 - rho_S + rho_S * rho_T )
for i = 1 : numSamples
    weights(i) = rho_S + ZT * weights(i) / ZT + ZT
% Adjust the sample weights to auxiliary layer
for i = 1 : numSamples
    weights2(i) = weights2(i) * spower(bata_T0_1, error_rate)
end

% print bata_T0
bata_T0
sum = 0
for i = 1 : numSamples
    left = logsumexp(
        labelCost(bata_T0 + rho_S * i, intexp((N - 2) / N))
        - labelCost(bata_T0 - rho_S * i, intexp((N - 2) / N))
    )
    right = 0.5 * logsumexp( / bata_T0, intexp((N - 2) / N))

    if left >= right
        predict(i) = 1
    else
        predict(i) = 0
    end
    sum = sum + 1
end
% print('predict')
% print('sum')
predict = sum ./ numSamples * predict
sum = sum / numSamples
predict = sum ./ numSamples * predict
return predict

end

% calculate F1 score, precision
total = numSamples
correct = numSamples * predict
total = sum(correct) / total, order = 2
[total, sampleCost, trans_data, test_data, test_data_P] =
    cif = readStructFromCifar10('data/cifar10.mat', 't')
[splitTrain, validation, num_images - num_images / 10] % percent done
train_data = train_data, train_label, sample_weights(1, 80)

```

[illegible]